

Recursive Time Complexity

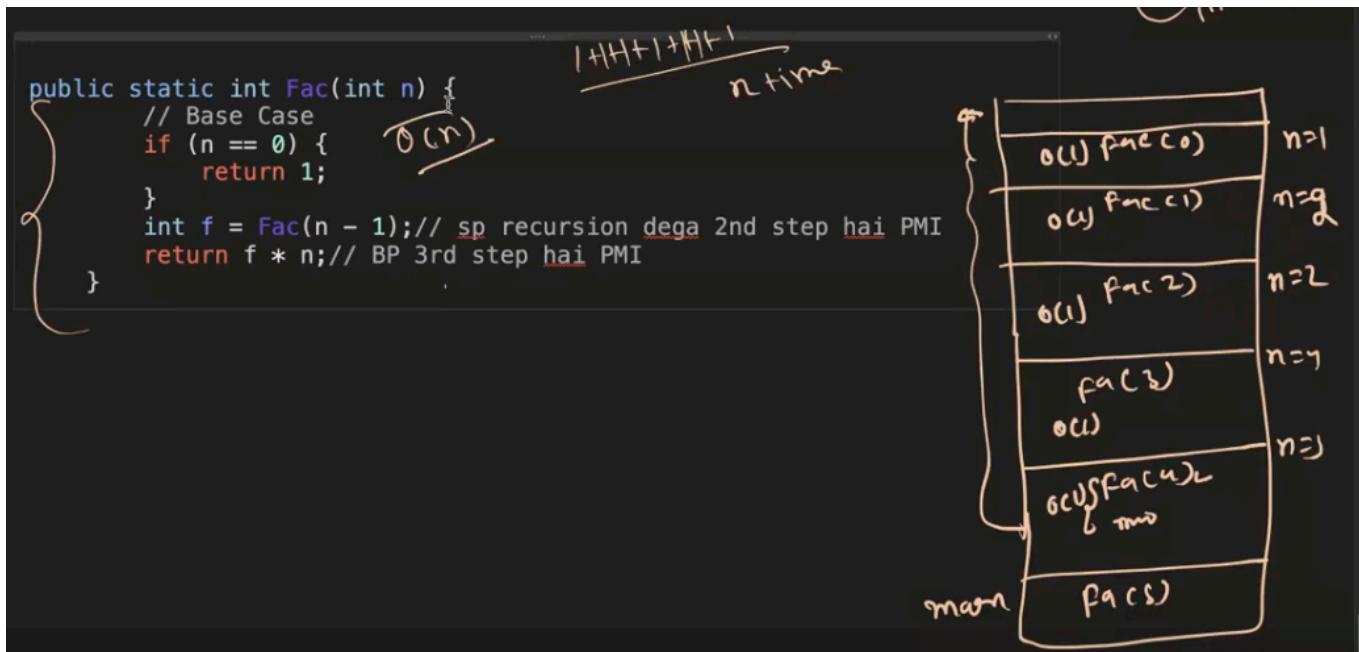
We will analyze the time complexity of recursion, focusing on two main cases:

1. When there is a single recursive call.
2. When there are multiple recursive calls.

When there is a single recursive call

Basically, to determine the time complexity of a recursive function, I look at how many recursive frames (or stack frames) are created — in other words, how many times the function calls itself. Then, I calculate the time complexity of each frame and sum them up. The total amount of work done across all recursive calls gives me the overall time complexity.

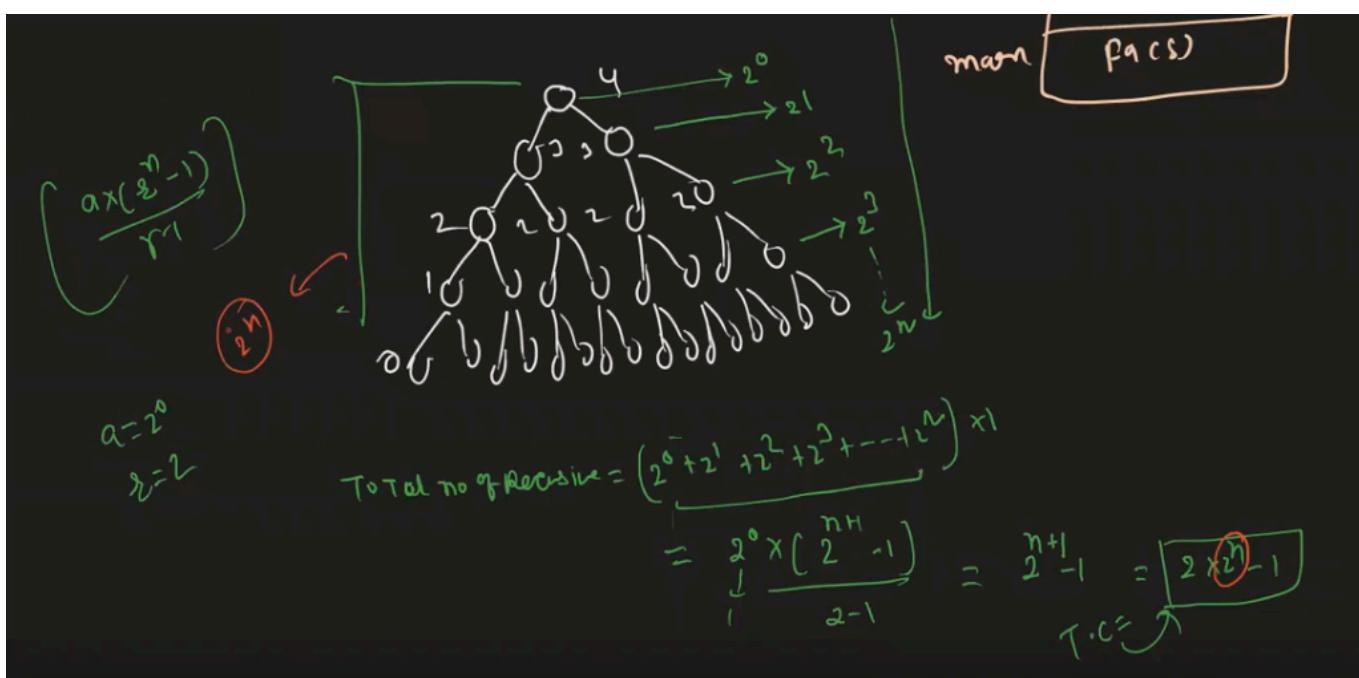
<https://github.dev/marvel2950/Java-DSA-1.0/tree/main/Day%2015>



When there are multiple recursive calls

Number of recursive calls to the power n.

<https://github.dev/marvel2950/Java-DSA-1.0/tree/main/Day%2016>



⌚ Power logN

```
public class Main {  
    public static void main(String[] args) {  
        int a = 3;  
        int n = 4;  
        System.out.println(pow(a, n));  
    }  
  
    public static int pow(int a, int n) {  
        if (n == 0) {  
            return 1;  
        }  
        int ans = pow(a, n / 2);  
        ans = ans * ans;  
        if (n % 2 != 0) {  
            ans = ans * a;  
        }  
        return ans;  
    }  
}
```

📘 What is a Recurrence Relation?

Function of time which helps in finding time complexity

💻 Recurrence Relation for Factorial

When analyzing the time complexity of computing the **factorial of n** , we consider how the function behaves recursively.

To compute **factorial(n)**, we must first compute **factorial($n - 1$)**. This means the time taken to compute **factorial(n)** depends on the time taken to compute **factorial($n - 1$)**, plus a constant amount of work (typically the multiplication and function call).

So, the **recurrence relation** becomes:

$$T(n) = T(n - 1) + 1$$

This indicates that each recursive step takes constant time (**+1**), and the problem size reduces by 1 with each step, until the base case is reached (usually **$T(1)$** or **$T(0)$**).

```

public static int Fac(int n) {
    // Base Case
    if (n == 0) {
        return 1;
    }
    int f = Fac(n - 1); // sp recursion dega 2nd step hai PMI
    return f * n; // BP 3rd step hai PMI
}

```

$T(n)$

\rightarrow Time function \rightarrow Time complexity help

$$T(n) = T(n-1) + 1$$

Expansion Method (for Solving Recurrence Relations)

Factorial

$$\{ T(n) = T(n-1) + 1 \}$$

BASE CASE :- input size is $\neq 1$

$$\begin{aligned}
T(n) &= T(n-1) + 1 \\
&= [T(n-2) + 1] + 1 \\
&= T(n-2) + 2 & n-1 < 1 \\
&= [T(n-3) + 1] + 2 & k = n - 1 \\
&= T(n-3) + 3
\end{aligned}$$

| → after K expansion

$$T(n) = T(1) + (n-1)$$

$$\begin{cases} T(n) = c + T(n-1) \\ T(n) = O(n) \end{cases}$$

Binary Search

$$\boxed{T(n) = T\left(\frac{n}{2}\right) + 1}$$

Base case :-

input size is 1.

$$\begin{aligned} T(n) &= T(n/2) + 1 \\ &= T(n/4) + 1 + 1 \\ &= T(n/8) + 1 + 1 + 1 \end{aligned}$$

{ → after K
expansion }

$$T(n) = T(n(2^K)) + K$$

$$\frac{n}{2^K} = 1 \Rightarrow K = \log_2 n$$

$$T(n) = T(1) + \log_2 n$$

$$T(n) = C + \log_2 n$$

$$\boxed{T(n) = O(\log_2 n)}$$

Merge Sort

$$\boxed{T(n) = 2 \cdot T\left(\frac{n}{2}\right) + n}$$

Base case :-

input size is 1

$$T(n) = 2T(n/2) + n$$

$$\frac{n}{2^K} = 1 \Rightarrow K = \log_2 n$$

$$= 2[2T(n/4) + n/2] + n$$

$$= 4T(n/4) + n + n$$

$$= 4[2T(n/8) + n/4] + 2n$$

$$= 8T(n/8) + 3n$$

{ → after K
expansion }

$$T(n) = 2^K T(n(2^K)) + K \cdot n$$

$$\begin{aligned} T(n) &= 2^{\log_2 n} \cdot T(1) \\ &\quad + \log_2 n \cdot n \end{aligned}$$

$$\begin{aligned} T(n) &= n \cdot T(1) + \\ &\quad n \log_2 n \end{aligned}$$

$$\boxed{T(n) = O(n \log n)}$$

Fibonacci

$$\{ T(n) = T(n-1) + T(n-2) + 1 \}$$

$$T(n) = [T(n-2) + T(n-3) + 1] + [T(n-3) + T(n-4) + 1] + 1$$

relation keeps growing → X

$$T(n-1) + T(n-2) + 1 \leq T(n-1) + T(n-1) + 1$$

↑ calculate TC.

$$T(n) = 2 \cdot T(n-1) + 1$$

$$= 2[2 \cdot T(n-2) + 1] + 1$$

$$= 2^2 \cdot T(n-2) + 2$$

$$= 2^2 \cdot [2 \cdot T(n-3) + 1] + 2$$

$$= 2^3 \cdot T(n-3) + 3 \cdot 2$$

↓ after K
expansion &

$$T(n) = 2^K \cdot T(n-K) + K$$

Base case :-

input size is 1

$$\begin{cases} n-1 <= 1 \\ K = n \end{cases}$$

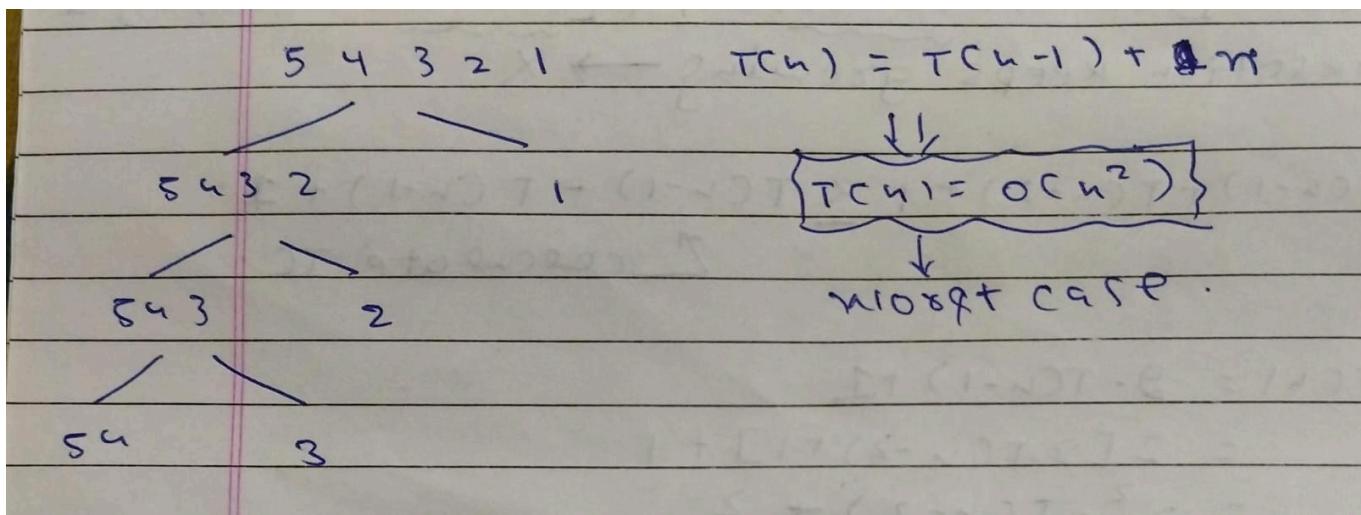
$$T(n) = 2^n \cdot T(1) + n$$

$$T(n) = 2^n \cdot C + n$$

$$\boxed{T(n) = O(2^n)}$$

What is Randomized Quick Sort?

Randomized Quick Sort is a variation of the classic Quick Sort algorithm where the **pivot element is chosen randomly** instead of using a fixed position (like the first or last element).



Why Randomize the Pivot?

- **Classic Quick Sort** can perform poorly (worst-case $O(n^2)$) if the pivot is consistently the smallest or largest element, especially on already sorted or nearly sorted data.
- **Randomizing the pivot** reduces the chance of consistently poor splits.
- This leads to **better average-case performance** and makes the algorithm's runtime more predictable and closer to the average $O(n\log n)$.

How Randomized Quick Sort Works:

1. Choose a pivot randomly from the current array/subarray.
2. Partition the array into two parts:
 - o Elements less than the pivot.
 - o Elements greater than or equal to the pivot.
3. Recursively apply Randomized Quick Sort to both parts.
4. Combine the results.

Time Complexity:

- **Average case:** $O(n\log n)$
- **Worst case:** $O(n^2)$, but with very low probability due to random pivot selection.

```

public class Main {
    public static void main(String[] args) {
        int[] arr = { 5, 7, 2, 1, 8, 3, 4 };
        sort(arr, 0, arr.length - 1);
        for (int i = 0; i < arr.length; i++) {
            System.out.print(arr[i] + " ");
        }
    }

    public static void sort(int[] arr, int si, int ei) {
        if (si >= ei) {
            return;
        }
        int idx = partition(arr, si, ei);
        sort(arr, si, idx - 1);
        sort(arr, idx + 1, ei);
    }

    public static int partition(int[] arr, int si, int ei) {
    }
}

```

```
Random rn = new Random();
int ri = rn.nextInt(ei - si) + si;

int tt = arr[ei];
arr[ei] = arr[ri];
arr[ri] = tt;

int item = arr[ei];
int idx = si;
for (int i = si; i < ei; i++) {
    if (arr[i] <= item) {
        int temp = arr[i];
        arr[i] = arr[idx];
        arr[idx] = temp;
        idx++;
    }
}
int temp = arr[ei];
arr[ei] = arr[idx];
arr[idx] = temp;
return idx;
}

}
```